

# K-Nearest Neighbors

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July 29, 2025

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**High values of  $K$**  will result in smoother decision boundaries  
 $\implies$  lower variance but also higher bias

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What is the time complexity of finding  $k$  nearest neighbors naively?



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- ▶ **Scalability:** Approximate methods needed for large datasets